Day 1: Introduction to Machine Learning ARISE 2020: ECE Machine Learning Lab

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- 1 Introductions
- 2 What is Machine Learning
- 3 Course Outlin
- 4 Setting Up Pytho
- 5 Problem Solving as an Engineer
- 6 Basics of Programming in Pythor
- 7 Matrices and Vectors
- 8 Visualizing Data
- 9 Lab: Plotting Functions
- ML vs. Al, Why the Hype Today?
- Supervised Learning
- 11 Supervised Learning



Tell us about yourself

- Write down the following pieces of information in your book:
 - Name
 - Grade
 - Where are you from?
 - What time did you wake up this morning (exact times please!)?
 - What is your favorite thing to do during Quarantine?
 - What do you want to get out of this class?
 - One rule you'd like to propose for the classroom
- Discuss what you've written down
- We'll visualize this data in python later today
 - Link to excel sheet here



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Machine Learning

- Most recent exciting technology
- We use these algorithms dozens of times a day
 - Web search engine
 - Face detection
- Machine Learning is an important component to achieve the big AI dream
- Practice and critical thinking are the key to learn machine learning



Definition

■ Machine Learning is a field of study that gives computers the ability to learn without being explicitly programmed.





Example: Digit Recognition

7 7 7 7 7 7 7 7 7 7

- Challenges with expert approach
 - Simple expert rule breaks down in practice
 - Difficult to translate our knowledge into code
- Machine Learning approach
 - Learned systems do very well on image recognition problems

```
def classify(image):
    ...
    nv = count_vert_lines(image)
    nh = count_horiz_lines(image)
    ...
    if (nv == 1) and (nh == 1):
        digit = 7
    ...
    return digit
```



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Machine Learning Problem Pipeline

- 1 Gather data
- 2 Visualize the data
- 3 Formulate ML problem
 - Regression vs Classification
 - Choose an appropriate cost function
- 4 Design the model and train to find the optimal parameters of the model
 - Prepare a design matrix
 - Perform feature engineering
 - Validate your choice of hyper-parameters using a cross-validation set
- 5 Evaluate the model on a test set
 - If the performance is not satisfactory, go back to step 4



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Course Outline

- What is Machine Learning and why is it so relevant today?
- Linear Regression
- Logistic Regression
- Back propagation algorithm
- Mini-Project on topics covered so far
- Neural Networks
- Conv Nets
- PyTorch tutorial
- Applications of CNN's and RNNs
- Self-Driving Cars Final Project



Course Format, Website, Resources

- Course Website: github.com/ApoorvaNandini/ARISE2020
 - Github: share collections of documents, repositories of code
 - Contains lecture slides, code notebooks, and datasets
 - Slides posted before lecture, demo code and solutions posted after
- Strongly encourage programming in Python via Google Colab
 - No installation required
- Will give additional resources based on student interest



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Setting Up Python

- Scripting Python
 - Sequentially execute lines of code
- Jupyter Notebook
 - Code Blocks and Text Blocks (markdown)
 - Variables saved in workspace (across code blocks)
- Google Colab
 - Jupyter Notebook online (no real-time collaboration)
 - No installation
 - Free GPU for 12 hours
- Set up 1 of these 3 things and print('hello world!')



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- **111** Supervised Learning



Black-Box Abstraction

[j+-¿]

- Can deal with many things with only knowing their **inputs** and **outputs**
- Need not know the details of the insides of our "boxes"
- Examples:
 - MTA turnstile
 - Python's print() function
 - Your Smart-Phone
 - Others?
- When does Black-Box Abstraction fail?



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Python Basics

- Program
 - We write operations to be executed on variables
- Variables
 - Referencing and interacting with items in the program
- If-Statements
 - Conditionally execute lines of code
- Functions
 - Reuse lines of code at any time



Python Basics

- Lists
 - Store an ordered collection of data
- Loops
 - Conditionally re-execute code
- Strings
 - Words and sentences are treated as lists of characters
- Classes (advanced)
 - Making your own data-type. Functions and variables made to be associated with it too.



Modules/Libraries/Packages

- NumPy: math, vectors and matrices
- MatPlotLib: plotting graphs, visualizing data
- Pandas: convenient for storing and retrieving data
- Sklearn: convenient wrapper for simple ML problems
- TensorFlow: computational graph, neural networks
- Keras: convenient wrapper for TF
- PyTorch: deep learning framework, used mostly by Research community



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Vectors

- A **vector** is an n-tuple of numbers or symbols
 - **E**x: integer 4-tuple: $\mathbf{v} = (1, 5, 2, 9)$
- Vectors of the same size may be added together, element-wise

■ Ex:
$$\mathbf{u} = (3, 6, 0, -5)$$

 $\mathbf{v} + \mathbf{u} = (1 + 3, 5 + 6, 2 + 0, 9 + (-5)) = (4, 11, 2, 4)$

- Vectors may be scaled by a number, element-wise
 - **E**x: $3\mathbf{v} = (3 \times 1, 3 \times 5, 3 \times 2, 3 \times 9) = (3, 15, 6, 27)$
- **Dot-Product**: sum of element-wise products of two vectors
 - Ex:

$$\mathbf{u} \cdot \mathbf{v} = 3 \times 1 + 6 \times 5 + 0 \times 2 + (-5) \times 9 = 3 + 30 - 45 = -12$$

- Length of a Vector (L2 Norm): $||\mathbf{x}|| = \sqrt{\mathbf{x} \cdot \mathbf{x}}$
 - **E**x: $||\mathbf{v}|| = \sqrt{1 + 25 + 4 + 81} = \sqrt{111}$
- Dimension of a vector
 - Ex: $\mathbf{u} = (3, 6, 0, -5)$; $\dim(\mathbf{u}) = 4$



■ A matrix is a rectangular array of numbers or symbols arranged in rows and columns

■ Ex: 2 by 3 matrix,
$$M = \begin{bmatrix} 1 & 9 & -12 \\ 15 & -2 & 0 \end{bmatrix}$$

■ Matrices of the same shape may be added together, element-wise

■ Ex:
$$A = \begin{bmatrix} 1 & 1 \\ 2 & 1 \end{bmatrix}$$
, $B = \begin{bmatrix} 0 & 8 \\ 7 & 11 \end{bmatrix}$, $A + B = \begin{bmatrix} 1 & 9 \\ 9 & 12 \end{bmatrix}$

■ Matrices may be scaled, element-wise

■ Ex:
$$aB = \begin{vmatrix} 0 & 8a \\ 7a & 11a \end{vmatrix}$$
, where a is a scalar



More About Matrices

- Matrices may be multiplied together provided their shapes meet the criteria

 - Criteria: for AB to be valid, # cols of A must equal the # rows of B
 - Result is a matrix with shape (# rows A, # cols B)
 - **EXECUTE** Ex: A is (2×3) and B is (3×5) , C = AB is (2×5)
 - Example on board
- **Transpose**: A^T swaps the rows and columns of matrix A
- Inverse: A^{-1} satisfies the equation $AA^{-1} = A^{-1}A = \mathbb{I}$
 - Square matrices only!



Vectors and Matrices

- We may consider a vector as a matrix
 - **Row Vector**: shape $(1 \times N)$ Ex: $\mathbf{v} = \begin{bmatrix} 1 & 5 & 2 & 9 \end{bmatrix}$

■ Column Vector: shape (N x 1)

Ex:
$$\mathbf{v} = \begin{bmatrix} 1 \\ 5 \\ 2 \\ 9 \end{bmatrix} = \begin{bmatrix} 1 & 5 & 2 & 9 \end{bmatrix}^T$$

- We'll consider vectors as column vectors by default
- Vector-Matrix Multiplication: on board
 - In what "dimension" is our vector starting, and where is it going to?
- Dot Product:



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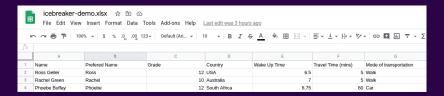


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Looking at our ice-breaker data in spreadsheets

- Columns have labels in the first row
- Collected data (numbers, words) follow below
- Let's export it to a Comma-Separated Values (CSV) file and open it





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Lab: Plotting Functions

- Generate and plot the following functions in Python:
 - Scatter plot of points: (0,1), (2,3), (5,2), (4,1)
 - Straight Line: y = mx + b
 - Sine-wave y = sin(x)
 - Polynomial e.g. $y = x^3 + 2$
 - Exponential e.g. $y = e^{-2x}$
 - Gaussian (Use $\sigma = 0.5$)
 - Choose a function of your own
- Create separate plots for each of the functions, Compute the mean and variance of each function
- Use Wikipedia and Numpy Documentation to search for mathematical formulas and python functions



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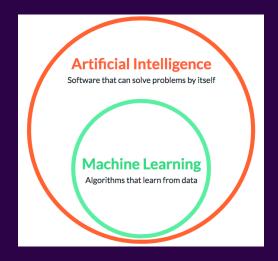


Artificial Intelligence

- Search
- Reasoning and Problem Solving
- Knowledge Representation
- Planning
- Learning
- Perception
- Natural Language Processing
- Motion and Manipulation
- Social and General Intelligence



Machine Learning





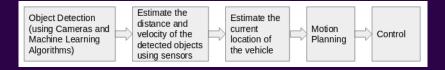
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Autonomous vs. Automated





Autonomous Example: Driver-less Cars





Why is Machine Learning so Prevalent?

- Database mining
- Medical records
- Computational biology
- Engineering
- Recommendation systems
- Understanding the human brain



Why Now?

- Big Data
 - Massive storage. Large data centers
 - Massive connectivity
 - Sources of data from internet and elsewhere
- Computational advances
 - Distributed machines, clusters
 - GPUs and hardware



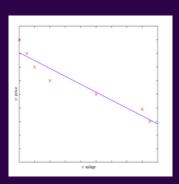
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What is Regression?

- Target variable is continuous-valued
- Example
 - \blacksquare Predict y = price of a car
 - From x = mileage, size, horsepower
 - Can use multiple predictors
- Assume some form of mapping
 - Ex: Linear mapping: $y = b + w_1x$
 - Find parameter b, w_1 from data
- Use target-feature pairings as examples to form model





What is Classification?

- Determine what class a target belongs to based on its features
- Example:
 - Predict y = what type of object is in a photo
 - From x = the pixels of the image
- Learn a model/function from features to target
- Use target-feature pairings as examples to form model





■ **Problem 1:** Categorizing credit card applications into those who have good credit, bad credit and those who fall in the gray area.



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■ Problem 4: Estimating change in climate.



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 - Regression Problem



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 - Target Variable: Predicting future temperatures



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- **Problem 4:** Estimating change in climate.
 - Regression Problem
 - Target Variable: Predicting future temperatures
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 - Classification Problem
 - Class labels: Normal Speech, Hate Speech
- **Problem 6:** Forecasting the energy demand in a region.



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- **Problem 6:** Forecasting the energy demand in a region.
 - Regression Problem
 - Target Variable: Predicting the amount of energy needed in the future



Thank You!

■ Next Class: ...

